## 6 Supplementary materials I

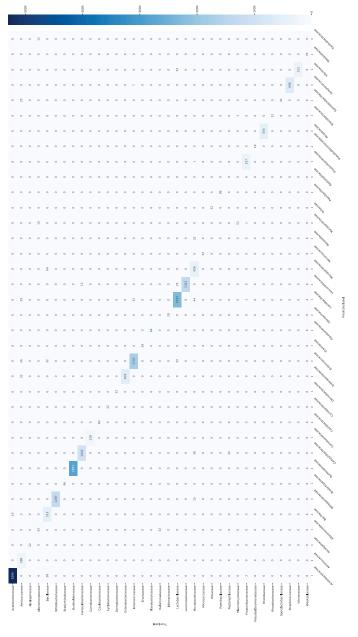


Figure 1: Confusion matrix on family level for flat KNN obtained in the novel strains scenario. We observed confusion for families *Aerococcaceae*, *Bacilaceae*, *Lactobacillaceae* and *Microbacteri*-

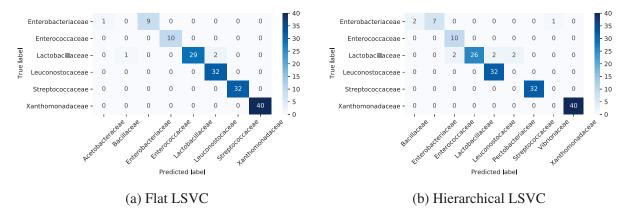


Figure 2: Confusion matrices on family level for LSVC obtained in the novel biological replicates scenario. Similar as in the novel strains scenario, we observed confusion for *Lactobacillaceae*.

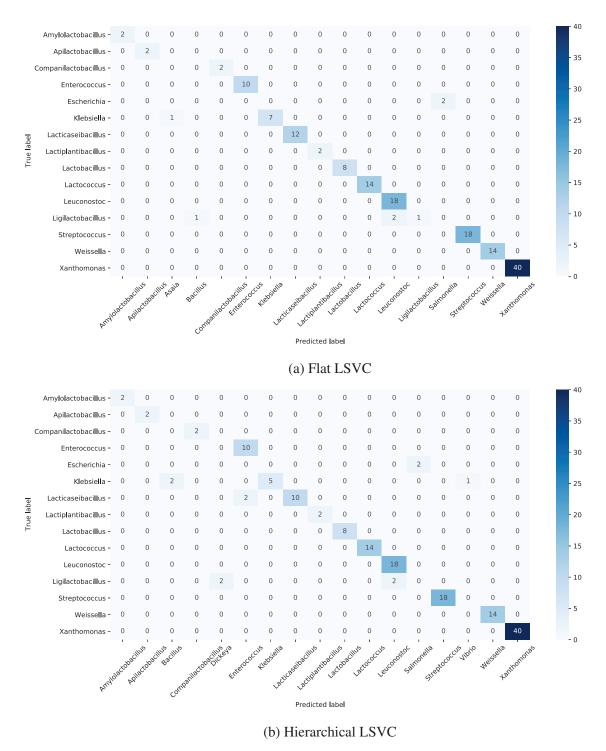


Figure 3: Confusion matrices on genus level for LSVC obtained in the novel biological replicates scenario.

Table 6: Overview of the different scenarios considered in the additional experiments, together with corresponding datasets. Statistics for the datasets are shown in Table 7.

SCENARIO	TRAIN	TEST
NOVEL STRAINS NOVEL SPECIES		$\begin{aligned} &GD_{test2,ID} \\ &GD_{test2,ID} \cup GD_{test2,OOD} \end{aligned}$

Table 7: Summary statistics for the different datasets used in the additional experiments  $(N - \text{number of spectra}, K_f - \text{number of unique families}, K_g - \text{number of unique genera}, K_{sp.} - \text{number of unique species}, K_{st.} - \text{number of unique strains}, ID - \text{in-distribution}, OOD - \text{out-of-distribution}).$ 

DATASET	N	$K_{\mathrm{f}}$	$K_{g}$	$K_{\mathrm{sp.}}$	$K_{ m st.}$
GD <sub>train2</sub>	79338	91	240	1035	2131
GD <sub>test2,ID</sub>	147	22	41	110	139
$GD_{test2,OOD}$	53	24	34	52	52

## 7 Supplementary materials II

As an additional experiment, we compared different classifiers, excluding logistic regression due to low performance in our work, with the ClinProTool of Bruker Daltonik GmbH & Co. KG (Germany, http://www.bruker.com/) for the novel strains and novel species scenario. Moreover, we constructed a test set ( $GD_{test2}$ ) by taking a random sample of 200 spectra from the global dataset. Subsequently, for the remaining part of the global dataset, we constructed a training set ( $GD_{train2}$ ) by removing strains that are also present in the test set. Furthermore, the test set consists of two mutually exclusive sets: one which includes species observed during training ( $GD_{test2,ID}$ ) and one which includes species that were not observed during training ( $GD_{test2,OOD}$ ), respectively. The former test set is used to evaluate the novel strain identification scenario, whereas the union of former and latter is used to evaluate the novel species identification scenario – see Table 6 for an overview. Summary statistics for the different datasets are presented in Table 7.

For the Bruker software, we classify a spectrum by predicting the label of the top-1, in terms of log-score (i.e., similarity score), spectrum in the training set. When it comes to the novel species scenario, the reciprocals of the log-scores obtained by Bruker are used in the same spirit as the total uncertainty in Eqn. 5 is used by KNN and 1DQNN. When a sample from an unobserved species is analyzed by the Bruker software, one might expect a lower similarity score with the training set. Results obtained for the different scenarios are presented in Table 8 and Table 9. For both scenarios, it is clear that the Bruker software is outperformed by most of the machine learning models discussed in this work.

Table 8: Additional results for novel strains scenario. Accuracies are reported on phylogenetic levels: (F)amily, (G)enus and (S)pecies. For each machine learning model, we report the performance for flat and hierarchical classification. H-XXX denotes the hierarchical classification implementation of method XXX and Bruker denotes the ClinProTool of Bruker Daltonik GmbH & Co. KG (Germany, http://www.bruker.com/).

MODEL	$\mathrm{Acc}_F$	$\mathrm{Acc}_G$	$Acc_S$
LSVC	0.9184	0.8980	0.8027
H-LSVC	0.9388	0.9184	0.8095
RF	0.8639	0.8435	0.7211
H-RF	0.9116	0.8980	0.7551
KNN	0.9388	0.9320	0.8095
H-KNN	0.9388	0.9320	0.8095
1DCNN	0.9184	0.8980	0.7347
H-1DCNN	0.9252	0.9116	0.7551
Bruker	0.9252	0.8639	0.7551

Table 9: Additional results for novel species scenario. Area under the ROC curve (AUROC) and area under the precision-recall curve (AUPR) are reported for out-of-distribution detection based on total uncertainty for KNN and 1DCNN, and the reciprocal of the top log-score obtained by the ClinProTool of Bruker Daltonik GmbH & Co. KG (Germany, http://www.bruker.com/) for Bruker.

MODEL	AUROC	AUPR
KNN	0.6024	0.5165
1DCNN(0.2)	0.8634	0.8634
1DCNN(0.4)	0.8647	0.6259
1DCNN(0.6)	0.8588	0.6340
1DCNN(0.8)	0.8734	0.6827
Bruker	0.8312	0.5065

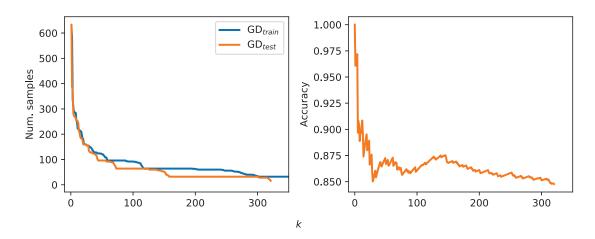


Figure 4: Left plot: number of samples (y-axis) plotted for the k most frequently observed species (x-axis) in the train  $(GD_{train}, blue)$  and test set  $(GD_{test}, orange)$  for the novel strains scenario (see Table 1). Right plot: accuracy obtained for KNN in the novel strains scenario (y-axis) when considering the k most frequent observed species (x-axis) in the test set  $(GD_{test})$ . A maximum accuracy is obtained when only looking at the most frequent species (i.e., k=1), while the lowest accuracy is obtained when including all species in the test set (i.e., k=321). From this plot, it is clear that the performance for the novel strains scenario increases when excluding the least represented species from the analysis.